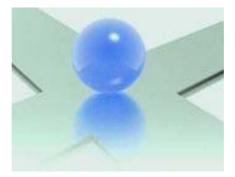
No-Failure Design and Disaster Recovery

Lessons from Fukushima

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Israel Institute of Technology



 $lectures \talks \lib \no-fail-disas-rec01.tex \quad 3.12.2014$

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1 *Highlights*

§ Major nuclear reactor accidents occur.

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- § Innovation dilemma.
- § No-fail design vs disaster recovery capability.

2 Lessons of Fukushima: No-Failure Design and Disaster Recovery

 $[\]label{eq:lectures} talks lib fukushima-lesson 01.tex 26.11.2014$



Figure 1: Int'l Nuclear Event Scale. (Wikipedia)

- § Nuclear plant accidents:
 - Major (INES 7):
 - Fukushima, Japan 11.3.2011.
 - Chernobyl, Ukraine, 26.4.1986.



Figure 2: Int'l Nuclear Event Scale. (Wikipedia)

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 - Major (INES 7):
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 - Chernobyl, Ukraine, 26.4.1986.
 - Serious (INES 6):
 - Kyshtym, USSR, 29.9.1957.
 - With wider consequences (INES 5):
 - \circ Windscale fire, UK, 10.10.1957
 - 3 Mile Island, Harrisburg, PA, 28.3.1979.
 - Lucens partial core meltdown (Switzerland), 21.1.1969
 - Others.

§ Approx nuclear power statistics (Aug 2011):

- 432 reactors in 30 countries (ENS).[‡]
- 366GWe installed capacity (ENS).
- 14,570 reactor years of experience (ENS).

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 $^{^{\}dagger}$ This is probably a substantial under-estimate. The numerator is too small: 3 reactors were seriously damaged, not 1. The denominator is too large: we should only take reactor-years from industrial democracies.

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= prob. of no major accident in 1 calendar year.

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= 33 year recurrence time for INES 7.

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- § Is 33 year recurrence long or short?

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§ Are there more lessons or challenges from Fukushima?

3 Science-Based Modeling

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- § We must understand the modern attitude: strengths and limitations.

Fundamental Physics Maxwell's Equations



Figure 4: James Clerk Maxwell, 1831–1879.

$$\nabla \cdot E = \frac{\rho}{\varepsilon} \qquad \nabla \cdot B = 0$$
$$\nabla \times E = -\frac{\partial B}{\partial t} \qquad \nabla \times B = \mu J + \mu \varepsilon \frac{\partial E}{\partial t}$$

- Positivism: From basic science to technology: Radio, X-ray diagnosis, CAT scan, wifi, remote sensing,
- Engineering education: sciences not crafts.

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Empirical Physics

Finite Element Modeling

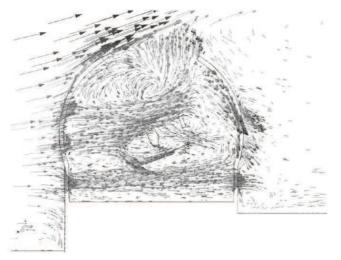


Figure 7: Example of velocity field on the vertical mid section

Figure 5: Velocity field around a structure.[‡]

• If we know the physics

we can

calculate anything.

• Methodology: simulation vs experiment.

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[‡]R. Codina, C. Morton, E. Oñate and O. Soto, http://www.cimne.com/eo/publicaciones/files/PI181.pdf

Computational Social Science

Econometric Modeling

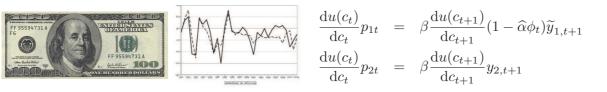


Figure 6: \$100, US GDP growth,[‡] Lucas asset pricing model.

• From the dry science

to

policy formulation.

• Methodology: social engineering.

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[‡] Saul H. Hymans, Forecasting and Econometric Models, The Concise Encyclopedia of Economics, http://www.econlib.org/library/Enc/ForecastingandEconometricModels.html

Computational Megalomania?

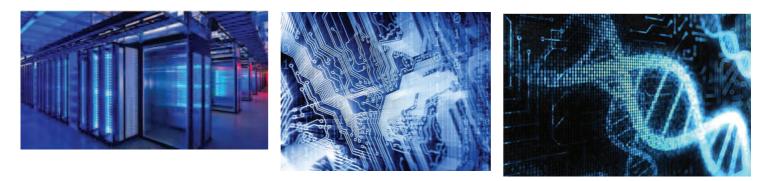


Figure 7: Computers and their aspirations.

• If you can't measure it, it's not real (logical positivism).

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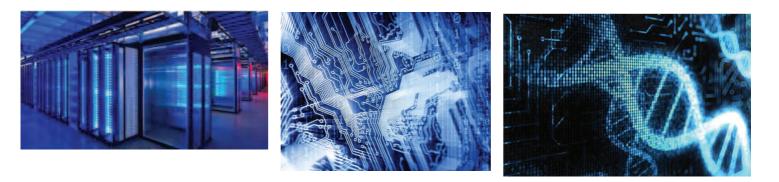


Figure 8: Computers and their aspirations.

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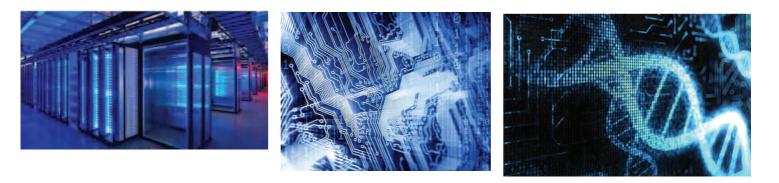


Figure 9: Computers and their aspirations.

- If you can't measure it, it's not real (logical positivism).
- If it's not a number, it's not important. (What about meaning?)
- We can compute anything. (Archimedes' modern lever?)

- § Quantitative model types:
 - Fundamental physics.
 - Empirical physics.
 - Computational social science.
 - Computational megalomania?

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§ The challenge:

Uncertainty, surprise, ignorance, change. Info-gaps.

4 Info-Gap Uncertainty: Examples

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$\sim \sim Thames \ Flood \ Barrier \sim \sim$

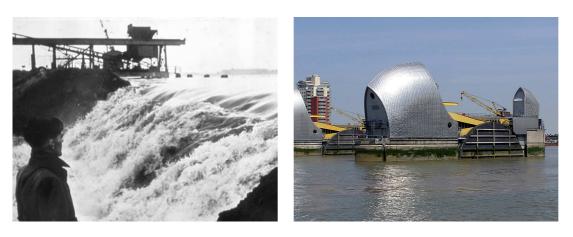


Figure 10: 1953 barrier breach. Figure 11: Barrier element.

§ Some facts:

- 1953: worst storm surge of century.
- Flood defences breached.
- 307 dead. Thousands evacuated.
- Canvey Island in Estuary devastated.
- Current barrier opened May 1984.

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§ Thames 2100:

Major re-design of flood defences.

§ Uncertainties:

- **Statistics** of surge height:
 - Fairly complete: most years since 1819.
 - Planning for 1000-year surge.
- Global warming: sea level rise.
- Tectonic settling of s. England.
- Damage vs flood depth.
- Human action: dredging, embanking.
- Urban development.
- § Severe Knightian uncertainties: Gaps in knowledge, understanding and goals.

$\sim \sim$ Fukushima Nuclear Reactor $\sim \sim$



Figure 12: Sea wall breach.



Figure 13: Hydrogen explosion.

§ Some facts:

- 11.3.2011: Richter-9 earthquake in NE Japan.
- Tsunami followed shortly.
- Sea wall breached: fig. 12.[‡]
- Hydrogen explosion several days later. Fig. 13.[‡]
- Slow disaster recovery.

§ Info-gaps:

- Sub-system interactions.
- Institutional constraints.

 $[\]label{eq:lictures} talks lib ig-unc01 fukushima.tex 17.7.2015$

 $[\]ddagger http://www.dailymail.co.uk/news/article-1388629/Japan-tsunami-destroyed-wall-designed-protect-Fukushima-nuclear-plant.html$

$\sim \sim Assay \ Spatially \ Random \ Material \sim \sim$





Figure 14: Nuclear Waste.

Figure 15: Gold Ore Vein.

- Detector type, location, number?
- Info-gaps:
 - \circ Spatial distribution of analyte.
 - \circ Spatial heterogeneity of matrix.

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$\sim\sim$ Interest rate after 9/11 $\sim\sim$

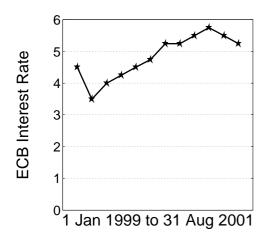


Figure 16: ECB Interest Rates

• Rate fairly constant through Aug 2001

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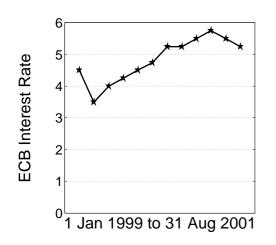




Figure 17: ECB Interest Rates

Figure 18: **11 Sept 2001.**

- Rate fairly constant through Aug 2001
- After 9/11 ECB will reduce the rate.
- Info-gap:
 - Reduce by how much?
 - What is ECB decision model?

$\sim \sim Climate \ Change \sim \sim$

§ The issue:

Sustained rise in green house gases results in temperature $r^{i^{s^e}}$ which results in adverse economic $im_{Pa_{c_t}}$.

§ Models:

- Temperature change: $\Delta \mathbf{CO}_2 \Longrightarrow \Delta T$.
- Economic impact: $\Delta T \Longrightarrow \Delta GDP$.
- § The problems:
 - Models highly uncertain.
 - Data controversial.

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§ E.g., IPCC model for

Uncertainty in Equil'm Clim. Sensi'ty, S.

- Likely range: 1.5° C to 4.5° C.
- Extreme values highly uncertain.
- 95th quantile of S in 10 studies:

Mean: 7.1^oC. St. Dev: 2.8^oC.

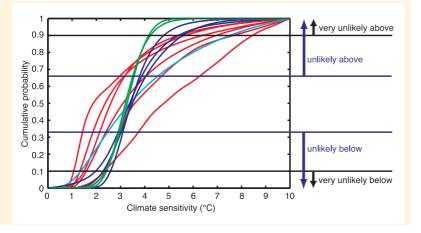


Figure 19: IPCC ch.10, p.799.

$\sim \sim Profiling Criminals \sim \sim$



Figure 20: Profiling raises arrests.

- Profiling: focus policing resources.
 - Arrests rise in profiled group.
 - Crime rises in other groups.
 - Everybody happy?
- Info-gaps: Uncertain response functions.

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$\sim\sim$ Summary $\sim\sim$

§ Severe Knightian uncertainties: Gaps in knowledge, understanding and goals.

§

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$\sim \sim Summary \sim \sim$

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- § Severe Knightian uncertainties: Gaps in knowledge, understanding and goals.
- § Info-Gap models of uncertainty:
 - Disparity between what is known and what needs to be known for responsible decision.
 - Unbounded family of sets of events (points, functions or sets).
 - No known worst case.
 - No funcs. of probability, plausibility, likelihood, etc.
 - Hybrid: info-gap model of probabilities.

5 Innovation Dilemma

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- § Choose between two options:
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 - Higher uncertainty because it's new.

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 - Option 2:
 - Standard. State of the art.
 - Lower uncertainty because it's well known.

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- Automotive collision control:
 - \circ Sensor-based computer control (innov).
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- Risk taking or avoiding:
 - \circ Nothing ventured, nothing gained (innov).
 - \circ Nothing ventured, nothing lost (SotA).

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 - \circ Use models to predict outcomes.
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 - \circ Specify wonderful outcome as piration.
 - \circ Use models to predict opportuneness.
 - \circ Choose best ops of wonderful outcome.

- § Decision strategies.
 - Outcome optimization:
 - \circ Use models to predict outcomes.
 - Choose predicted best option.
 - Max-min (maximize the min reward):
 - Specify level of uncertainty.
 - \circ Use models to predict worst outcomes.
 - \circ Choose the best worst-outcome.
 - Robust satisficing:
 - Specify critical outcome requirements.
 - \circ Use models to predict robustness.
 - Choose best rbs of adequate outcome.
 - Opportune windfalling:
 - \circ Specify wonderful outcome as piration.
 - \circ Use models to predict opportuneness.
 - \circ Choose best ops of wonderful outcome.

§ Question:

Which strategy suitable for innovation dilemma?

6 Thames Flood Barrier



Figure 21: 1953 barrier breach. Figure 22: Barrier element.

§ Some facts:

- 1953: worst storm surge of century.
- Flood defences breached.
- 307 dead. Thousands evacuated.
- Canvey Island in Estuary devastated.
- Current barrier opened May 1984.

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§ Thames 2100:

Major re-design of flood defences.

§ Uncertainties:

- **Statistics** of surge height:
 - Fairly complete: most years since 1819.
 - \circ Planning for 1000-year surge.
- Damage vs flood depth.
- Global warming: sea level rise.
- Human action: dredging, embanking.
- Urban development.
- Tectonic settling of s. England.

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Small probability of large damage.

§ Decision: choose a design.

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- § Decision: choose a design.
- § Challenge: Uncertainty.
 - Our data, understanding is limited.
 - Our goals may be unclear, conflicting.
- § Design strategy: Robust satisficing.
 - How wrong can we be, and the design is still adequate? (Satisficing.)
 - How large a surprise can the design tolerate? (Robustness.)

- Design 1:
 - Innovative technology.
 - Early warning system.
 - Adaptive channeling.

0

- Design 1:
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 - State of the art technology.
 - Solid dykes.
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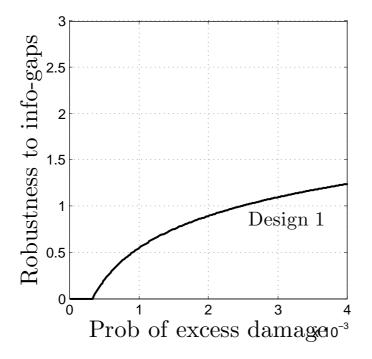
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- § Choose design 1? Design 2?
 - Responsible decision?
 - Robust to ignorance?

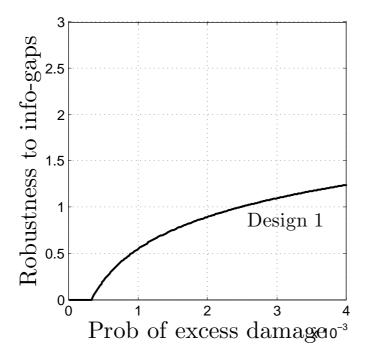
§ Robustness to info-gaps vs Probability of excess damage. Design 1



§ Trade off:

Less demanding outcome has greater robustness.

§ Robustness to info-gaps vs Probability of excess damage. Design 1



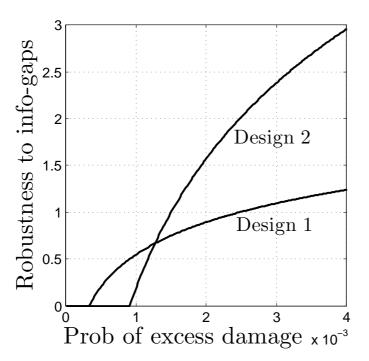
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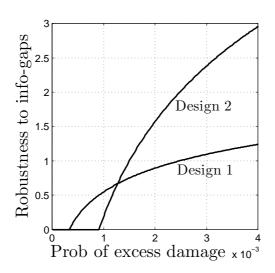
§ Zeroing:

Estimated outcome has zero robustness.

§ Comparing 2 designs.

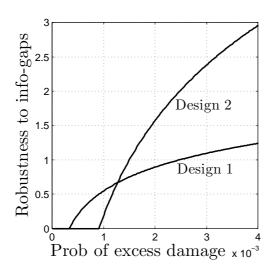


- § Design 1 (innov) estimated to be better. Zero robustness of estimates.
- § Design 2 (SotA) more robust for $P > P_{\times}$.
- § Innovation dilemma.



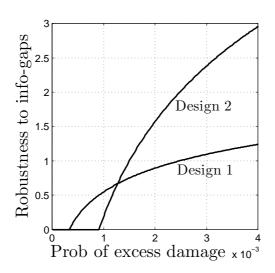
§ Outcome optimization:

- Find best models. (Maybe probability.)
- Predict best-outcome design.



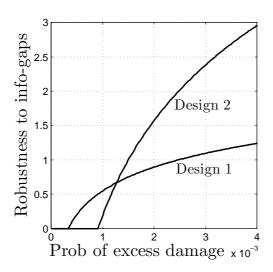
§ Outcome optimization:

- Find best models. (Maybe probability.)
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- § Robust-satisficing:
 - Identify critical outcome.
 - Maximize rbs of critical outcome.



§ Outcome optimization:

Des 1 predicted better than Des 2.

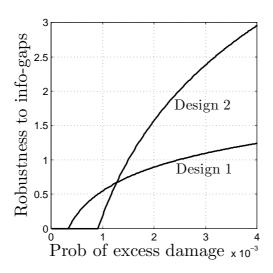


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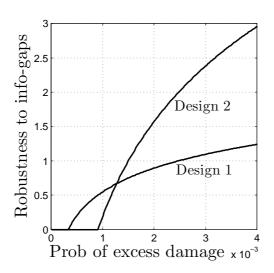
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Design 2 more robust for $P > P_{\times}$.

- § Resolve innovation dilemma:
 - Value judgment on outcome requirement.
 - Robustly satisfy requirement.



- § Robust-satisficing strategy: Robustly satisfy performance requirement.
- § Question:

Is robustness a good bet?

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Info-Gap Theory

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7 Conclusion

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- Prepare Disaster Recovery Capability.

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§ Closing question:

No-fail design and disaster recovery capability are both necessary for critical technology. How to decide the technology is feasible?